

Weekly Report(July.22.2019-July.28.2019)

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Abstract

This week I continue to train Resnets and get an average accuracy of 95%. And this mini project seems to come to an end.

1 Question from last week

- First, I run more different models to ensure it is not a coincidence. The result are shown at Table 1. The performances are relatively steady and still worse than SGD's. Therefore, the phenomenon I metioned last report still exists. It seems not a coincidence.

Table 1: momentum=0.95, epochs=100

| Learning Rate | Batch Size | Accuracy |
|---------------|------------|----------|
| 0.01 | 256 | 78.42% |
| 0.003 | 256 | 74.21% |
| 0.001 | 256 | 72.33% |
| 0.0003 | 256 | 63.17% |
| 0.0001 | 256 | 47.81% |

- The loss curves are shown in Figure 1. obviously when learning rate becomes smaller, model need more epochs to converge. Last report I didn't realize this since models optimized by SGD can all converge in 100 epochs. Sorry about this mistake.
- Give the model larger learning rate and it can indeed converge in 100 epochs with a good performance. The results are shown in Table 2 and loss curves are shown in Figure 2.

Table 2: momentum=0.95, epochs=100

| Learning Rate | Batch Size | Accuracy |
|---------------|------------|----------|
| 1 | 256 | 87.29% |
| 0.3 | 256 | 87.97% |
| 0.1 | 256 | 86.58% |
| 0.03 | 256 | 83.41% |

2 ResNet

Since there are several models in ResNets and train a model once needs much time, I just trained ResNet-18 with different hyper parameters and deeper ResNets with same hyper parameter. Hyper parameters and result of each model are shown as following:

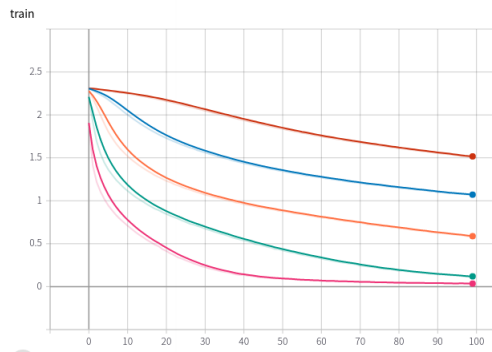


Figure 1: small learning rate

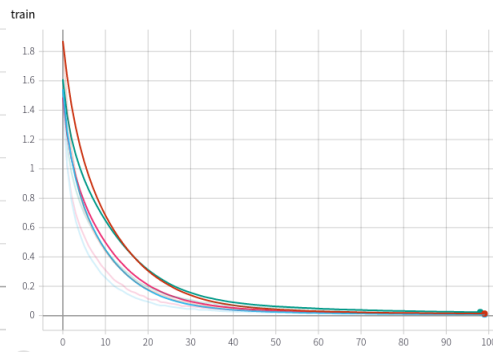


Figure 2: large learning rate

Figure 3: loss curves with Adadelta

- In this part, replace the fc layer for all nets with a new fc layer initialized with `nn.init.normal_(params, mean=0.0, std=0.1)`
- The accuracies of ResNet-18 are shown in table 3. Generally It performs well.
 - As learning rate becomes smaller, net performs better since loss can be more closer to minimum point.
 - Since larger batch size will run out of memory, small batch size is more used here. Within this scale, larger batch size works better.

Table 3: momentum=0.9, epochs=10

| Learning Rate | Batch Size | Accuracy |
|---------------|------------|----------|
| 0.01 | 8 | 86.18% |
| 0.01 | 16 | 91.59% |
| 0.01 | 32 | 92.99% |
| 0.003 | 8 | 94.93% |
| 0.003 | 16 | 95.68% |
| 0.003 | 32 | 95.53% |
| 0.001 | 8 | 92.15% |
| 0.001 | 16 | 95.56% |

- ResNet-34, ResNet-50, ResNet-101 are also trained with the hyper parameters performing best in ResNet-18. The Accuracies are shown in table 4 and their loss curves are shown in Figure 4.

Table 4: learning_rate0.003, batch_size=16, momentum=0.9, epochs=20

| Net | Accuracy |
|------------|----------|
| ResNet-18 | 95.68% |
| ResNet-34 | 95.32% |
| ResNet-50 | 95.45% |
| ResNet-101 | 96.23% |

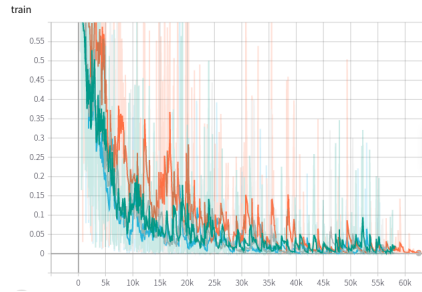


Figure 4: ResNets Loss Curves

3 Summary of Mini Project

Generally Mini project has come to an end. The best performances of these three models are shown as following:

- AlexNet: with learning rate=0.01, batch size=128, momentum=0.9, get accuracy of 88.09%.
- VGG Net: with learning rate=0.0003, batch size=8, momentum=0.9, get accuracy of 91.10%.
- ResNets: ResNet-101 with learning rate=0.003, batch size=16, momentum=0.9, get accuracy of 96.23%.

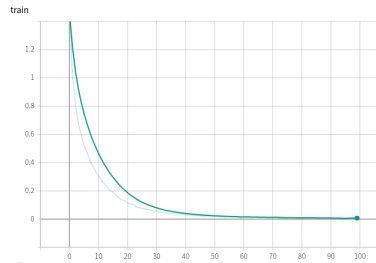


Figure 5: AlexNet Loss Curve

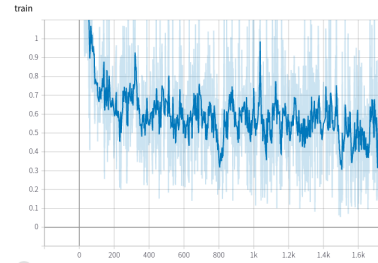


Figure 6: VGG Loss Curve

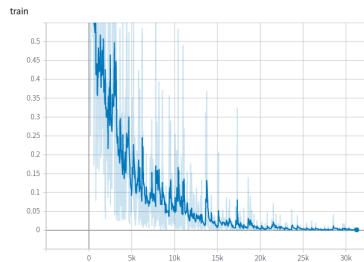


Figure 7: ResNet Loss Curve

Figure 8: loss curves of best performance